TEMPORAL DECOMPOSITION STUDIES OF GRB LIGHTCURVES

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Abstract. Gamma-ray bursts (GRB) are extremely energetic events and produce highly diverse light curves. Light curves are believed to be resulting from internal shocks reflecting the activities of the GRB central engine. Hence their temporal studies can potentially lead to the understanding of the GRB central engine and its evolution. The light curve variability time scale is an interesting parameter which most models attribute to a physical origin e.g., central engine activity, clumpy circum-burst medium, or relativistic turbulence. We develop a statistical method to estimate the GRB minimum variability time scale (MVT) for long and short GRBs detected by GBM. We find that the MVT of short bursts is distinctly shorter than that for long GRBs supporting the possibility of a more compact central engine of the former. We find that MVT estimated by this method is consistent with the shortest rise time of the fitted pulses. Hence we use the fitted pulse rise times to study the evolution of burst variability time scale. Variability time is in turn related to the minimum bulk Lorentz factor. Using this we relate the GRB spectral evolution to the evolution of the variability time scale.

1 Introduction

Gamma-ray bursts (GRBs) are short, intense and distant flashes of γ -rays that occur at random locations in the sky with their peak power in the 200-500 keV range. During their appearance, they often outshine all other sources in the γ -ray sky combined. The temporal structure of GRBs exhibits diverse morphologies. They can vary from a single smooth pulse to extremely complex light curves with many erratic pulses with different durations, amplitudes, and fine structures. Physically, several mechanisms have been invoked to interpret GRB temporal variability. The leading scenario is to attribute the light curve variability to the activity of the central engine (Rees & Mészáros, 1994, Sari & Piran, 1997). There are no direct observations of the central engine. Most of the bursts exhibit variability on time scales that are much shorter than the burst durations. According to internal shock model of GRBs, the γ -ray light curves result from collisions between shells with different values of the bulk Lorentz factor Γ . Within such a scenario, the observed light curves can be directly connected to the behavior of the central engine (Lei, 2007, Lu, Huang & Zhang, 2008).

One approach for probing light curves which has received attention (Norris $et\ al.$, 2005, Bhat $et\ al.$, 2012) is to express them as a series of displaced pulses, each with a parametric form. There is an appeal to this approach because fitting routines are well-understood and interpretations of rise time, decay time, full width at half max, etc., are possible.

GRB light curves exhibit variability on various time scales. In this paper we present a new statistical method of estimating the minimum value of such variability time scales of a GRB and relate it to the minimum value of the fitted pulse rise time. This in turn can be related to the minimum Lorentz factor of the relativistic shells emitted by the central engine. The evolution of the minimum Lorentz factor is then related to the spectral evolution of the GRBs.

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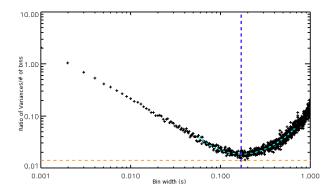


Fig. 1: Variation of the ratio of the variances per bin to the histogram bin-width. At very fine bin-widths the GRB signal is indistinguishable from background fluctuations and hence the ratio decreases monotonically with increasing bin-width. At larger bin-widths the signal is clearly visible from the background and hence the ratio per bin starts increasing. The turn over bin-width is defined as the minimum variability time scale where the bin-width is optimum. Cyan dashed line shows a fitted parabola around the minimum that has a minimum at a bin-width indicated by the vertical dashed line in blue. See text for more details.

2 Minimum Variability Time-scale in GRBs

GRB light curves are generally binned in to narrow time bins. Such light curves with high variability at low power may show variations which are not statistically significant. While statistically significant variability could become statistically insignificant at finer bin-widths and significant variability could vanish if the bin-widths are too coarse. By a comparison of the GRB prompt emission variability with the purely statistical variability of the background region we derive an optimum bin-width when the non-statistical variability in the light curve becomes significant.

We identify the prompt emission duration and an equal duration of background region. We then derive a differential of each light curve and compute the ratio of the variances of the GRB and the background. This ratio divided by the bin-width is plotted as a function of bin-width in figure 1. As can be seen at very fine bin-widths the ratio falls monotonically with increasing bin-width signifying that at such fine bin-widths the variations in the background and burst regions are statistically identical. In other words the signal in the burst light curve is indistinguishable from Poissonian fluctuations. Later at certain bin-width the variation starts deviating from the $\frac{1}{bin-width}$ behavior. We measure the bin-width at this valley by fitting a parabola and the bin-width at the minimum of the parabola is called the optimum bin-width t_b . t_b is also interpreted as the minimum variability time scale of the GRB t_v . At this bin-width the variability in the GRB light curve becomes detectable compared to the normal fluctuations seen in the background. We are therefore confident that the short-term variability is real and is not an artifact of data reduction or statistical fluctuations.

Another possibility is that t_b could be a function of the signal-to-noise ratio of the GRB rather than an intrinsic feature of the GRB light curve. For this we generated synthetic light curves of the same GRB by adding Poisson noise to the fitted lognormal pulses and adding them to the fitted background with added noise. The optimum bin-width was estimated for each of the synthetic light curve derived by varying the signal-to-noise ratio over a few orders of magnitude. It was found that the value of t_b is not a strong function of the signal-to-noise ratio.

Recently the minimum variability time scales were estimated for long and short GRBs by a model independent method based on a wavelet decomposition technique (MacLachlan et al., 2012). In figure 2 we compare the minimum variability time scales estimated by MacLachlan et al. (2012) with those estimated by the present method for the same GRBs. The blue line shows the ideal case when the two quantities are equal. The figure shows that the minimum variability time scales estimated by two different techniques are statistically consistent with each other. Hence we conclude that the present method indeed estimates the minimum variability time scale of a given GRB. MacLachlan et al. (2012) also demonstrate that the minimum variability time scale estimated by the wavelet decomposition technique is also consistent with the minimum of the rise-times of the fitted pulses using the lognormal shape for the individual pulses (Bhat et al., 2012).

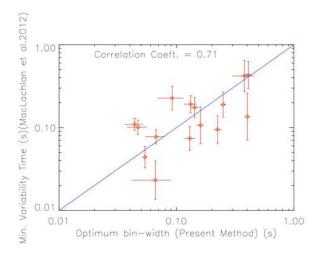


Fig. 2: A comparison of the minimum variability time scale estimated by McLachlan et al.(2012) with the optimum bin-width estimated by the present method. They seem to correlate well.

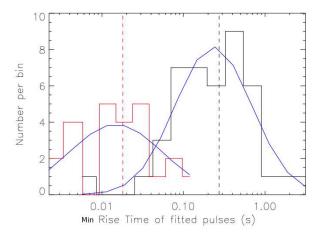


Fig. 3: A distribution of the minimum pulse rise time for long and short GRBs. The minimum variability time scale or the minimum fitted pulse rise time clearly shows a bimodal distribution showing that it can be a parameter to distinguish between long and short GRBs.

2.1 Minimum Variability Time-scale as a GRB type identifier

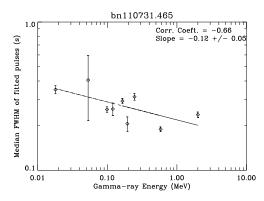
GRBs are generally classified as long or short depending on their duration T_{90} and their hardness ratios. Long GRBs ($T_{90} > 2s$) are generally soft and short GRBs ($T_{90} < 2s$) are generally hard. They are considered to form two different GRB classes (Kouveliotou, etal, 1993). The measurement of spectral lags is another tool in the study of GRBs and their classification since short GRBs exhibit negligible lags. However classification schemes based on any or all of these parameters result in significant overlap of GRBs of either type. Hence we need more such identifying parameters to uniquely identify a GRB type. Here we have another parameter, t_v , to add to that list. Figure 3 shows a distribution of minimum rise-times for long and short GRBs. The minimum pulse rise-times of short GRBs are distinctly shorter (at least by a factor of 15) than that of long GRBs. In other words the minimum variability time scale, t_v , can be used as another parameter to identify short GRBs.

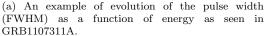
3 Minimum Variability Time scale and minimum Lorentz factor

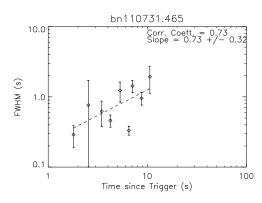
While t_v has been shown to be consistent with the minimum rise-time of fitted pulses, the rise-times of the fitted pulses to the prompt emission light curve can be used to trace the evolution of the variability time scale at any time during the GRB prompt emission. One can also derive a lower limit on the bulk Lorentz factor (Γ_{min}) given the variability timescales and observations of the highest energy photons at any time during a burst. Constraining the bulk Lorentz factor, Γ , of the jet is a major challenge in understanding the GRB physics, the mechanism for launching the jet as well as high-energy emission. High-energy γ -rays produced and emitted from the GRB jet are subject to $\gamma\gamma \to e^+e^-$ pair production process with soft target photons, and absorbed in situ. The interaction rate of this process and corresponding opacity, $\tau_{\gamma\gamma}$, for the high-energy γ -rays depends on the target photon density and can be significant when both the high-energy and target photons are produced in the same physical region. Highly relativistic motion, with a bulk Lorentz factor $\Gamma \gg 1$, of such an emission region can reduce the $\gamma \gamma$ interaction rate and $\tau_{\gamma \gamma}$ greatly by allowing for a larger emitting radius and a smaller target photon density. Observation of a γ -ray spectrum up to an energy E_{max} thus can be used to put a lower limit on Γ (Lithwick & Sari, 2001, Razzaque, Mészáros, & Zhang, 2004, Granot, Cohen-Tanuji, & do Couto e Silva, 2008, Ackermanetal 2010). Thus the evolution of the variability time scale leads to the study of the evolution of the bulk Lorentz factor which in turn is related to the evolution of the γ -ray opacity $\tau_{\gamma\gamma}$ during the prompt emission phase of a GRB. A delayed onset of the GeV photons, seen in several GRBs detected in the Fermi LAT, the emission is interpreted as due to the time evolution of the opacity in a GRB outflow (Hascoët et al. 2012). In addition, As pointed out by Granot et al. (2008), due to the temporal evolution of $\tau_{\gamma\gamma}$, the opacity cut-off in a time-integrated spectrum will be smoother than a sharp exponential decay: the cut-off transition will be close to a power-law steepening. This time evolution takes place within a given γ -ray pulse, and can be even stronger in a complex burst where the light curve is made up of many pulses (Aoi et al. 2010). Hence a study of the evolution of Γ_{min} would lead to a better understanding of the possible connection between the temporal structure of the light curve and the spectral evolution of the GRB. Figure 4a shows variation of the fitted pulse width (FWHM) as a function of γ -ray energy while figure 4b shows a variation of the FWHM as a function of time since the GRB trigger time. It is well known that for lognormal pulse shapes the rise-time and FWHM are strongly correlated (Bhat et al., 2012). Hence figure 4a demonstrates that the variability time scale decreases with increasing γ -ray energy while figure 4b shows that the variability time scale decreases since the trigger, indicating a trend like the hard-to-soft spectral evolution in GRBs. Figure 5 shows a typical example of the observed Γ_{min} evolution during a bright short Fermi GRB 090510(Abdo et al., 2009). Here the Γ_{min} is estimated at two different epochs of the GRB (0.6s-0.8s, left panel and 0.8s-0.9s, right panel, post trigger) by assuming that the highest energy of the γ -ray (E_{max}) emitted in each interval originate from the same physical region as the observed low energy photons in the same time interval. The t_v were approximated conservatively to the FWHM of the fitted pulse during each interval. The data points with error bars correspond to the Γ_{min} calculated for the best-fit $t_v = \text{FWHM}$ and FWHM/2 in the respective time intervals.

4 Summary

A method is developed to estimate the optimum bin-width of the light curve to carry out pulse decomposition analysis of GRBs. The optimum bin-width is interpreted as the minimum variability time scale of the GRB because it is found to be statistically consistent with that estimated by an independent method which in turn is found to be consistent with the lowest of the rise-times of the fitted pulses to deconvolve the entire light curve of the GRB. The variability time scales were also estimated using the fwhm of the fitted pulses during the course of the burst and study the evolution of the bulk Lorentz factor. The variability time scales, t_b , were estimated at different γ -ray energies by the same pulse fitting technique using GRB light curves in different energy ranges. Using the estimated bulk Lorentz factor one can test the location of γ -ray emission regions in the internal shock scenario. The γ -ray emission radius is given by $R \sim 2\Gamma^2 ct_b/(1+z)$. Thus an energy dependent variation of t_b can be interpreted as collisions of shells at different radii producing γ -rays of different energies.







(b) Evolution of the pulse width (FWHM) as a function of time since trigger as seen in GRB1107311A.

Fig. 4

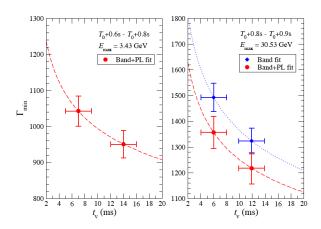


Fig. 5: The Γ_{\min} of the GRB 090510 prompt mission region as a function of the γ -ray variability time-scale t_v . The highest and low-energy (MeV) photons in each time-interval are assumed to originate from the same physical region. The data points correspond to the Γ_{\min} calculated for the best-fit $t_v = \text{FWHM}$ and FWHM/2 in the respective time intervals.

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